

Memoryless Thermodynamics? A Reply

Dibyendu Mandal,^{1,*} Alexander B. Boyd,^{2,†} and James P. Crutchfield^{2,‡}¹*Department of Physics, University of California, Berkeley, CA 94720, U.S.A.*²*Complexity Sciences Center and Physics Department,
University of California at Davis, One Shields Avenue, Davis, CA 95616*

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We reply to arXiv:1508.00203 ‘Comment on “Identifying Functional Thermodynamics in Autonomous Maxwellian Ratchets” (arXiv:1507.01537v2)’.

I. INTRODUCTION

Several years ago, Chris Jarzynski and one of us (DM) introduced a solvable model of a thermodynamic ratchet that leveraged information to convert thermal energy to work [1, 2]. Our hope was to give a new level of understanding of the Second Law of Thermodynamics and one of its longest-lived counterexamples—Maxwell’s Demon. As it reads in “bits” from an input string Y , a detailed-balance stochastic multistate controller raises or lowers a mass against gravity, writing “exhaust” bits to an output string Y' .

A complete understanding of the ratchet’s thermodynamics requires exactly accounting for all of the information embedded the input and output strings and how that information is changed by the ratchet. To simplify, we assumed the input bits came from a biased coin and so the input information could be measured using the *single-bit Shannon entropy* $H[Y_0]$. The information in the output string was much more challenging to quantify, since correlations are necessarily introduced by the action of the memoryful ratchet. Unfortunately, due to mathematical complications arising from this, we could only estimate the single-bit entropy $H[Y'_0]$ of the output. Which, it must be said, is only an upper bound on the actual information per output bit. Nonetheless, the estimate of the change $\Delta H = H[Y'_0] - H[Y_0]$ from input to output was good enough to show that the ratchet was quite functional, operating as an “engine” in some regimes and an “eraser” in others.

Following in this spirit, the three of us here recently introduced a similar memoryful ratchet for which all of the informational correlations in the output bit string can be calculated exactly and in closed form [3]. As a result, one of its contributions is that we could then show that the change $dh_\mu = h_\mu[Y'] - h_\mu[Y]$ in the *Shannon entropy rate* $h_\mu[X] = \lim_{\ell \rightarrow \infty} H[X_0 X_1 \dots X_\ell]/\ell$ allowed one to identify all of the ratchet’s thermodynamic functionality. We emphasized, in particular, that using single-bit Shannon entropy ΔH would miss much of that functionality, as $H[Y'_0] \geq h_\mu[Y']$. And, as such, we generalized Refs. [4, 5] single-bit ΔH “Second Law” to use the Shannon en-

trophy rate dh_μ . The underlying methods leveraged a new way to account for the information storage and transformation induced by memoryful channels [6]. A similarly complete analytical treatment of a companion Demon—Szilard’s Engine—was recently given by two of us (AB and JPC) [7].

II. SPECIAL CASE OF THE MEMORYLESS TRANSDUCER

A recent arXiv post [8] complained that our work [3] is misleading in certain aspects. It also claims priority over our entropy-rate Second Law [3, Eq. (4)], stating that Eq. (24) of Ref. [9] is the same. This is mathematically incorrect. Moreover, our Ref. [3] is very clear about its contributions. In short, our treatment is more general, since it considers the much broader class of Demons with arbitrary memory. Such Demons, as we describe in our manuscript, can be represented as memoryful channels, otherwise known as transducers [6]. In stark contrast, Ref. [9]’s treatment is sufficient only for describing memoryless channels; a highly restricted, markedly simpler case. More to the point, its methods are inapplicable to our memoryful channel setup. This error occurs in the proof of Ref. [9]’s Eq. (24) as it contains a statement that can be violated by memoryful channels. We provide two counterexamples to this erroneous statement in our response below. Finally, the case of memoryless Demons violates the spirit of Refs. [1, 2]’s original work. The mathematical errors and misinterpretation of physical relevance subvert the arXiv post’s claims. We now turn to respond to its three specific comments in greater detail.

Comment 1

In his first comment, the arXiv post’s author mentions that the following sentences in our paper give a “*very strong misleading impression that the paper above is the **first** to incorporate correlations successfully in general*” (using his own words). This is simple misreading, as our text makes clear:

We introduce a family of Maxwellian Demons for which correlations among information

* dibyendu.mandal@berkeley.edu

† abboyd@ucdavis.edu

‡ chaos@ucdavis.edu

bearing degrees of freedom can be calculated exactly and in compact analytical form. This allows one to precisely determine Demon functional thermodynamic operating regimes, when previous methods either misclassify or simply fail due to the approximations they invoke.

Note that this explicitly mentions the solvable aspect of our model—that correlations can be calculated exactly in a compact, analytical form. We stand by the claim that ours is the first such solvable model. We did not claim to be the first to consider correlations. More pointedly, the author’s actual article [9] *does not have any model with calculable correlations*. We justify the second sentence quoted above on identifying functional thermodynamics through explicit calculations and diagrams in Sec. V of our paper. The arXiv post ignores these.

The author claims that the “*main result in Section 4 of [1] was exactly the same as the above mentioned upper bound on the extracted work in terms of the change in the joint entropy ...*” In this, he refers to Eq. (4) of our paper and claims that he had derived it before as Eq. (24) of his paper [9]. While we agree that our equation superficially looks like the infinite-time limit of the author’s equation, their relationship is different than a glance suggests:

- The author’s proof of Eq. (24) [9] does not apply to our setup. This is because the author considered the much simpler case of memoryless channels, whereas we considered the much more mathematically challenging case of memoryful channels. (We return to this point again in context of the 3rd comment.)
- Appendix A in our paper clearly shows that Eq. (4) there is valid only in the asymptotic limit of stationary input bits for a finite-state Demon. (These are standard assumptions in the field.) In absence of these assumptions, we have a more general form of the Second Law discussed in detail in Appendix A [3]. The arXiv post neglects these discussions.

Comment 2

The arXiv post quotes the following from our paper:

In effect, they account for Demon information-processing by replacing the Shannon information of the components as a whole by the sum of the components’ individual Shannon informations. Since the latter is larger than the former [19], these analyses lead to weak bounds on the Demon performance.

And, then goes on to claim that the second assertion may not be true if the incoming bits $\{Y_i\}$ are correlated. This

is the case in the author’s Ref. [9], where the sum of individual entropy differences is actually stronger, under the additional assumption that the Demon is memoryless. We agree. But, as the author himself points out, our claim is true if the incoming bits are uncorrelated. We explicitly state that we are considering this case, where the input is uncorrelated, in the paragraph following Eq. (4). And, this happens to be the case for all the exactly solvable models of Maxwell’s Demon developed so far (referred to by “they” in the above quote). (We reiterate, the author has not given any exactly solvable model of Maxwell’s Demon with calculable correlations in [9].)

The author did not sufficiently consider the remainder of our development before expressing his criticism in public. After Eq. (4), we explicitly mention the sufficient condition of uncorrelated incoming bits for Eq. (4) to be stronger than Eq. (2).

According to the author “*the point in second law and its extensions ... should be to provide, first and foremost, an extended version of the second law in a faithful manner, namely, to show the increase of the real entropy of the entire system, including that of the information reservoir. In the correlated case, the latter is given by the change in the joint entropy of the symbols, regardless of whether or not this is smaller or larger than the sum of individual entropy differences.*” We disagree. This is nothing more than an attempt to rewrite the history of physics.

The primary emphasis of the Second Law from its very inception has been on the strongest possible bounds. When Sadi Carnot formulated the Second Law, it was **all** about maximum efficiency of heat engines—the maximum possible work that can be extracted [10]. Entropy was a derived concept, entering through the works of Clausius and Thompson [11].

The author mentions that “*bounds are useful when they are easier to calculate than the real quantity of interest, which is not quite the case in this context. Quite the contrary, joint entropies (especially of long blocks) are much harder to calculate.*” He fails to notice that we attained precisely this “hard” task by calculating exactly the entropy rate $h_\mu[Y'] = \lim_{\ell \rightarrow \infty} H[Y'_{0:\ell}]/\ell$. (We might, at this point, recommend the review of correlations and information in random-variable blocks presented by Ref. [12].) And, the entropy rate is smaller than the individual entropy difference, which in our case has **observable consequences**, as discussed in detail in Sec. V of our paper. Even the later part of his comment “*work itself ... depends only on the input and output marginals*” is not true in a generic memoryful situation. We have explicit examples (unpublished) where the extracted work also depends on correlations.

Comment 3

Here, the author claims that the “*bound in [57] ... is exactly the same as in eq. (4) of 1507.01537v2, except*

that in [57], no limit on is taken over the normalized entropies (but this is because even stationarity is not assumed there, so the limit might not exist). Moreover, while it is true that in the model of [57] the channel was memoryless, the derivation itself of this very same bound (in Section 4 of [57]) was not sensitive to the channel memorylessness assumption.” (Citation [57] corresponds to Ref. [9] here.) We agree that Eq. (4) in our paper appeared in a somewhat different form than in his paper, as Eq. (24). This is moot, however. His derivation does not apply to our case nor to the original solvable Maxwell’s demon [1]. In his justification, the author says that the “crucial step in [57] ... was the equality

$$H(Y'_i|Y_1, \dots, Y_{i-1}, Y'_1, \dots, Y'_{i-1}) = H(Y'_i|Y_1, \dots, Y_{i-1}), \quad (1)$$

which is the case when

$$Y'_i \rightarrow (Y_1, \dots, Y_{i-1}) \rightarrow (Y'_1, \dots, Y'_{i-1}) \quad (2)$$

forms a Markov chain, and this happens not only for a memoryless channel, but for any causal channel without feedback, namely,

$$P(Y'_1, \dots, Y'_n|Y_1, \dots, Y_n) = \prod_{i=1}^n P(Y'_i|Y_1, \dots, Y_i). \quad (3)$$

In physical terms, this actually means full generality.”

This analysis is incorrect. Equation (1) above is not sufficiently general, since it does not consider the case in which the Demon is a memoryful channel. When the Demon has memory, its internal state can depend on both the input past Y_1, \dots, Y_{i-1} and output past Y'_1, \dots, Y'_{i-1} . The Demon’s internal states store information about the past of Y or Y' and can communicate it to the outgoing bits of Y' . The author’s assertion of “full generality” is false. In fact, the memoryless assumption is violated for the original solvable model of Maxwell’s Demon [1] in which the Demon has three internal states. For a memoryful Demon Eq. (2) above is not a Markov chain. See Ref. [6]’s discussion of memoryful transduction.

To see how Eq. (1) can be violated, consider the case of a memoryful Demon that simply ignores the input bits and outputs a period-2 process. This means that there are two possible output words:

$$\Pr(Y'_1 Y'_2 \dots = 010101\dots) = \Pr(Y'_1 Y'_2 \dots = 101010\dots) = 1/2.$$

In this case the uncertainty of the i th output given the history of inputs is $H[Y'_i|Y_{1:i}] = 1$, since we are completely uncertain as to whether or not the i th bit is a zero or one. Note that we used the notational shorthand $Y_{1:i}$ to represent the random variables Y_1, Y_2, \dots, Y_{i-1} . When we also condition on the history of output bits, we find that we are completely certain of the next bit, since we know the output’s phase, and $H[Y'_i|Y_{1:i}, Y'_{1:i}] = 0$. The most general relation for the uncertainty of the output

is:

$$H[Y'_i|Y_{1:i}, Y'_{1:i}] \leq H[Y'_i|Y_{1:i}].$$

This inequality, replacing Eq. (1) above, renders the proof in the author’s paper inapplicable to our situation.

This reflects the fact that a memoryful ratchet can and typically does create correlations among the outgoing bits even though the incoming bits may not be correlated. In fact, we can exactly calculate the uncertainty in the next output bit conditioned on the infinite length input and output histories of the memoryful ratchet we describe in our Ref. [3]. When the ratchet is driven by a fair coin input process, the two quantities of interest are:

$$\lim_{i \rightarrow \infty} H[Y'_i|Y_{1:i}] = \frac{1}{2} \left(H\left(\frac{p}{2}\right) + H\left(\frac{q}{2}\right) \right),$$

where $H(b)$ is the binary entropy function for a coin of bias b [13], and:

$$\lim_{i \rightarrow \infty} H[Y'_i|Y_{1:i}, Y'_{1:i}] = \frac{1}{4} (H(p) + H(q)),$$

and their difference:

$$\begin{aligned} & \lim_{i \rightarrow \infty} (H[Y'_i|Y_{1:i}] - H[Y'_i|Y_{1:i}, Y'_{1:i}]) \\ &= \frac{1}{2} \left(H\left(\frac{p}{2}\right) - \frac{H(p)}{2} \right) + \frac{1}{2} \left(H\left(\frac{q}{2}\right) - \frac{H(q)}{2} \right) \\ &\geq 0, \end{aligned}$$

by the concavity of $H(\cdot)$. This is only zero when $p=q=0$. Thus, the assumption made in Eq. (1) is not just insufficiently general, but it is explicitly violated in the physical memoryful ratchet considered in our work.

On a more conceptual level, Eq. (4) in our paper is valid only in the asymptotic limit of a stationary input with a finite-state Demon. Otherwise, there would be natural generalizations incorporating the Demon’s entropy and its correlations with the bits, as is amply discussed in Appendix A of our paper.

III. SUMMING UP

As our response to the arXiv post’s Comment 3 just made plain, the essential issue reduces to the post’s author misapplying results for memoryless channels. Most directly, the post’s claim to priority for our entropy-rate Second Law is invalid. Perhaps the simple memoryless channel case, one very broadly adopted in elementary information theory [13], prevented the post’s author from appreciating this and related technical points. Whatever the motivation, it led to the post’s public airing of a series of grievances—grievances that derive not from misleading text, but from the author’s misinterpretation. That said, we do appreciate the opportunity to emphasize the central role of memory and structure in thermodynamics.

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